

Lecture 10

Plan: 1) Discuss pset /
~~misc. remarks~~;

BFS
→ S } Carathéodory's
Theorem.

2) Total unimodularity

#3) b) show if G factor
critical, Edmonds terminates
at single vertex.

Strategy: show ODD at end $\{^{\text{using}}\}$ (a).
of Edmonds is empty
b/c G_{final} is factor critical.

Easier: ODD is TB minimizer
for G_{original} ! From class.

$\Rightarrow \text{ODD} = \emptyset$ from part (a).

\Rightarrow implies G_{final} is singleton,
because G_{original} is connected.

Quiz see Canvas
announcement.

1) format: 2 hrs, 22 hr
window

for starting it, window starts

11:00 am Thurs Apr 1.

2) open note, no collaborators,
no internet access from canvas.

3) is practice quit already,
is assignment in canvas.

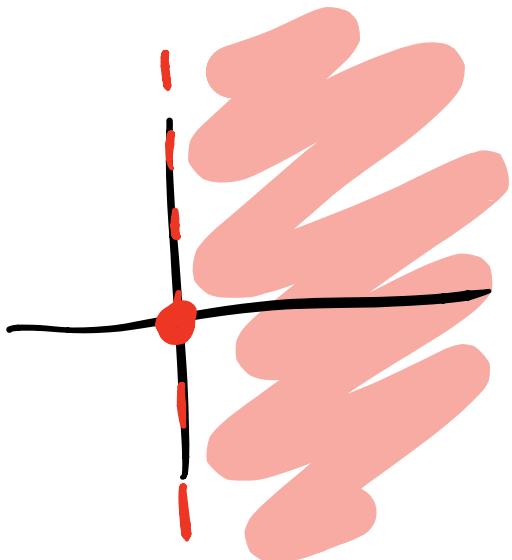
questions 3-4.

No lecture Apr 1.

#5) $\text{cone}(P) = \{0\} \cup \{x : \exists \lambda \text{ s.t. } \lambda x \in P\}$.

$$= \{0\} \cup \{x : \exists \vec{\lambda}^0 \text{ s.t. } A \vec{\lambda} x \leq b\}$$

$$= \{0\} \cup \{x : \exists \vec{\lambda}^0 \text{ s.t. } Ax \leq \lambda b\}.$$



cone could be
 $\{(x, y) : x > 0\}$

$\cup \{0\}$.

$$P = \{x : \exists \vec{\lambda}^0 \text{ s.t. } Ax \leq \lambda b\}$$

as projection of \tilde{P}

$$\tilde{P} = \{(x, \lambda) : Ax - b \leq 0, \lambda \geq 0\}$$

$$P = \tilde{P}_{n+1}$$

$$\tilde{A} = \left[\begin{array}{c|c} x & \lambda \\ \hline A & -b \\ \hline 0 & -1 \end{array} \right]$$

$$\tilde{P} = \{(x, \lambda) : \tilde{A}(x, \lambda) \leq 0\}$$

$$< 0$$

$$\tilde{A} = \left[\begin{array}{c|c} A & b \\ \hline 0 & -1 \end{array} \right]$$

$\left\{ \begin{array}{l} I_< \\ I= \\ I_> \end{array} \right.$ $\rightarrow b_i > 0$
 $\left\{ \begin{array}{l} I_< \\ I= \\ I_> \end{array} \right.$ $\rightarrow b_i = 0$.
 $\left\{ \begin{array}{l} I_< \\ I= \\ I_> \end{array} \right.$ $\rightarrow b_i < 0$

e.g can combine $I >$.
and final inequality $-\lambda < 0$.

$$a_i^T x - b_i \lambda \leq 0 \quad b_i < 0$$

$$+ (-b_i)(-\lambda < 0)$$

$$\rightarrow \boxed{a_i^T x < 0 \quad \forall i : b_i < 0}$$

Misc remarks

$\text{aff}(x)$

$\text{conv}(x)$

$\text{cone}(x)$

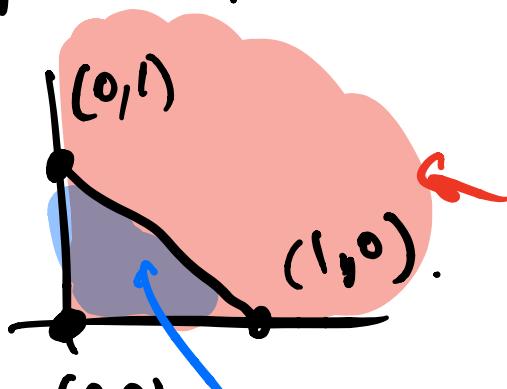
$$\text{aff}(x) = \{x_i \lambda_i : \sum \lambda_i = 1\}$$

$$\text{cone}(x) = \{x_i \lambda_i : \lambda_i \geq 0\}$$

$$\text{conv}(x) = \{x_i \lambda_i : \lambda_i \geq 0, \sum \lambda_i = 1\}.$$

is $\text{conv}(X) = \text{cone}(X) \cap \text{aff}(X)$?

NOT TRUE



convex hull.

affine hull all of \mathbb{R}^2

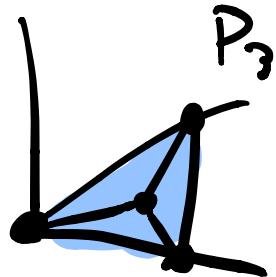
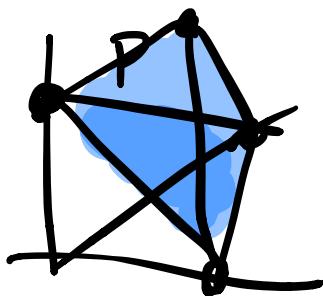
conic hull

affine \cap
 $\text{cone} =$
conic!

why not contradiction??

Misc. Remark #2

$P = \{x : Ax \leq b\}$ consider projection P_n



Know how
to write $P_n = \{x : Ax \leq \tilde{b}\}$
from Fourier - Motzkin.

what about if

$$P = \text{conv}(x_1, \dots, x_t)$$

$$P_n = \text{conv}((x_1)_n, \dots, (x_t)_n)$$

\uparrow
 get rid of
 final coordinate.

$$\text{vertices of } P_n \subseteq \{(x_1)_n, \dots, \dots, (x_t)_n\}.$$

$$P = \{x : Ax \leq b\}$$

↓

E.X.

add slack:

$$P = \{x : Ax \leq b, x \geq 0\}$$

$$Q = \{(x, s) : Ax + Is = b, x \geq 0, s \geq 0\}.$$

P is projection of Q.

Refresher: BFS

$$P = \{x : Ax \leq b, x \geq 0\}.$$

Recall: vertices of P

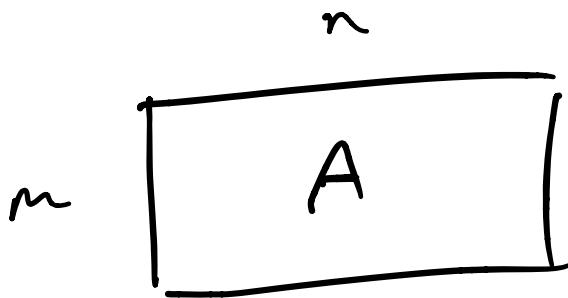
are basic feasible

- 0 + 1 - 1 -

Solutions \rightarrow i.e.

feasible solns obtained as follows:

- Remove redundant rows from A (so that $\text{rank } A = m$)



- Choose m columns B of A that form a basis for \mathbb{R}^m

A diagram of matrix A with m columns. The first m columns are highlighted with a bracket below them and labeled 'B'. The remaining columns are grouped together with a bracket below them and labeled ' A_B '.

$$\begin{bmatrix} & & & \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & b \\ 0 & 0 & 0 & b \end{bmatrix} = \begin{bmatrix} b \\ b \end{bmatrix}$$

x_B

- Solve $A_B X_B = b$,
set

$$x^* = \begin{bmatrix} 0 \\ X_B \\ 0 \end{bmatrix} \quad \left. \begin{array}{l} \text{---} \\ \text{---} \end{array} \right\} \in \mathbb{R}^3$$

we "proved"
in lecture
7 or 8.

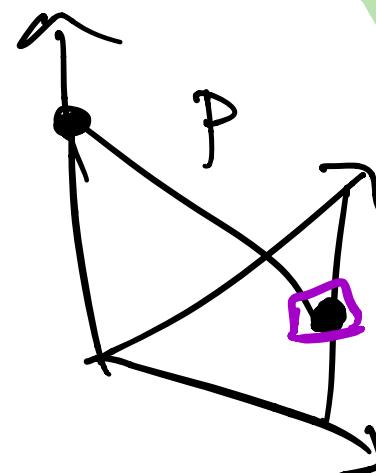
if $x^* \in P$, i.e. x^* feasible,
 x^* is bfs.

All vertices of P are bfs.

E.g.: in \mathbb{R}^3

$$P = \{(x_1, x_2) \in \mathbb{R}^3 : x_1 > 0, x_2 > 0, x_1 + x_2 \leq 1\}$$

$$\begin{bmatrix} 1 & -1 & 1 \\ 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ z \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$



3 potential bfs:

1) $z=0, \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

$$\Rightarrow (x, y, z) = \boxed{\left(\frac{1}{2}, \frac{1}{2}, 0\right)} \checkmark \text{ feasible,}$$

2) $y = 0, \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ z \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

$$\Rightarrow (x, y, z) = \boxed{(0, 0, 1)} \checkmark \text{ feasible}$$

3) $x = 0, \begin{pmatrix} 1 & 1 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} y \\ z \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

$$\Rightarrow (x, y, z) = \boxed{(0, 0, 1)} \checkmark \text{ feasible.}$$

Another example:

Carathéodory's

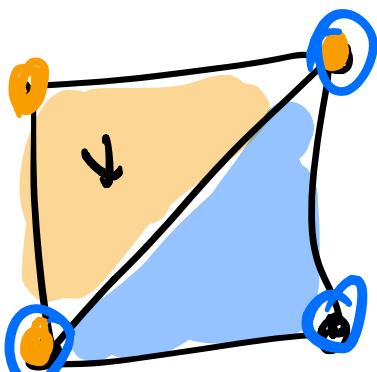
Theorem

a polytope

Every point in $P \subseteq \mathbb{R}^n$

can be written as a
convex combination of
 $\leq n+1$ vertices of P .

(holds for any
bounded convex
closed set
vertices \Rightarrow extreme
pts)



Proof

Let $x \in P$, $P = \text{conv}(v_1, \dots, v_t)$.

- consider ways to write x as convex combo of vertices v_1, \dots, v_t of P .
- Assume affine hull of vertices of P is \mathbb{R}^n (else could translate, rotate P to be $\subseteq \mathbb{R}^{n'}$ for $n' < n$.)

$$Q = \left\{ \lambda: \begin{array}{l} \sum_i \lambda_i v_i = x \\ \sum_i \lambda_i = 1 \\ \lambda_i \geq 0 \end{array} \right\}$$

all ways x
 is convex
 combo. $n+1$

$$= \left\{ \lambda: \underbrace{\begin{bmatrix} v_1 & \cdots & v_t \\ 1 & \cdots & 1 \end{bmatrix}}_{\text{matrix}} \lambda = x, \lambda \geq 0 \right\}.$$

$$\underbrace{\begin{pmatrix} \cdot & \cdot & \cdots & \cdot \end{pmatrix}}_A$$

* $\text{aff}(v_1, \dots, v_t) = \mathbb{R}^n \Leftrightarrow \text{rank } A = n+1$

(no redundant rows.).

Can use λ^* where λ^* is vertex;
So can take $\lambda^* = \underline{\text{BFS}}$?

λ^* has all but $n+1$ coordinates = 0!



Affine hull: X $\text{aff}(X)$ smallest
affine space containing it
affine spaces are translations of subspaces.



Total unimodularity

- Consider discrete set $X \subseteq \mathbb{R}^n$.

E.g.

- $X \in \mathbb{R}^{n \times n}$ incidence vectors of matchings
- $X \in \mathbb{R}^{n \times n}$ incidence vectors of independent (a.k.a. stable sets) in graphs.

- To optimize lineal functions over X , enough to do so

over $\text{conv}(X)$.

- For this, want simple polyhedral description

$$\text{conv}(X) = \{x : Ax \leq b\}.$$

ii
P.

Given proposed A, b ,
how to prove $\text{conv}(X) = P$?

- Easy to show $\text{conv}(X) \subseteq P$;

just check $Ax \leq b$ for all $x \in X$.

- How about $P \subseteq \text{conv}(X)$?

Harder!

- One way is Algorithmically:

Enough to show $\forall c \in \mathbb{R}^n$

$$\max_{x \in X} c^T x = \cdot \quad \leftarrow \text{primal}$$

By weak duality, enough to exhibit dual feasible y and $x \in X$ with



e.g. what we did for
min-weight perfect matching.
(MWPM).

- Today, another way:

show extreme points of P
integral.

E.g.

helps when you know

$$X = \{x \in \mathbb{R}^m : x \in P\}$$

i.e.

$X = \{\text{feasible set for}$
 $\text{some integer program.}\}$

$P \subseteq \text{conv}(X)$.

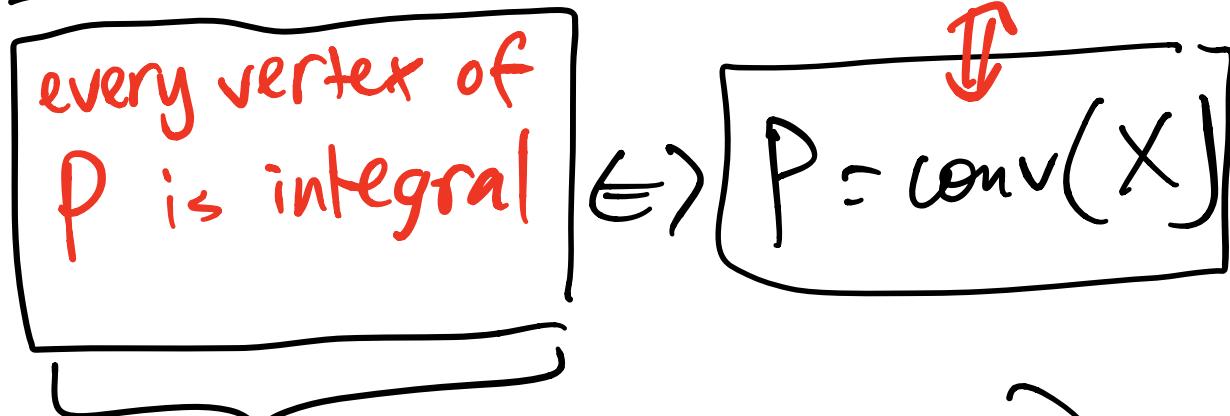
$P = \text{conv}(\text{vertices of } P)$

if vertices of
 P in \mathbb{Z}^m , then
they are in X .

Integer program

$$\begin{aligned} & \max c^T x \\ & x \in P \\ & x \in \mathbb{Z}^m \end{aligned}$$

In this case: $L.P. = I.P.$



if this happens, say P

often not the
case $\begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}$

Method of showing

this!

$$\begin{aligned}x+y &= 1 \\x+z &= 1 \\y+z &= 1\end{aligned}$$

Total Unimodularity

This is true when matrix A

is very special.

Def: matrix A is
totally unimodular (TU)

if every square submatrix
has determinant $+1, -1, 0$

e.g.

$$\begin{bmatrix} -1 & -1 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ not } \begin{bmatrix} 2 & 0 \\ 1 & 1 \\ -1 & \end{bmatrix}$$

Important because:

Theorem : (TU theorem)

Suppose A totally unimodular.

Then \forall integral b,

$$P = \{x : Ax \leq b, x \geq 0\}$$

is integral.

Pre-proof remarks:

- Same proof shows this also holds for

$$P = \{x : Ax \geq b, x \geq 0\}$$

or

$$P = \{x : Ax = b, x \geq 0\}.$$

- Converse: if $P = \{x : Ax \leq b, x \geq 0\}$ is integral for all integral b then A is TU.
 (but converse not true for $\{x : Ax = b, x \geq 0\}$).
-

Proof: First, reduce to equality by adding slack:

let

$$Q = \{(x, s) : Ax + Is = b, x \geq 0, s \geq 0\}.$$

Ex: Q integral $\Leftrightarrow P$ integral.

Fact: $\tilde{A} = [A | I]$ TU
 $\Leftrightarrow A$ TU.

e.g.

$$\det \begin{array}{|ccc|} \hline & a & b \\ & c & d \\ & e & f \\ \hline & 0 & 0 \\ & 0 & 1 \\ \hline \end{array} = \det \begin{pmatrix} a & b \\ c & d \end{pmatrix}.$$

In general: submatrix of \tilde{A}
is cols of A + cols of I ;
expand down cols I . to get
determinant of square A submatrix.

Thus: Forget about \tilde{A} , assume

$$P = \{x : Ax = b, x \geq 0\}.$$

where A is TU.

Recall: vertices are BFS!

$$\sim \begin{array}{c} n \\ \boxed{\begin{array}{|ccc|} \hline & 0 & AB \\ & AB & 0 \\ \hline \end{array}} \end{array} \begin{array}{c} \left[\begin{array}{c} 0 \\ x_B \\ 0 \end{array} \right] \\ = [b] \end{array} .$$

*

Claim: BFS integral.

has full rank

Why?

$$A_B X_B = b$$

we must
remove
redund.
rows

$$\Rightarrow X_B = A_B^{-1} b$$

But

$$X_B = \frac{1}{\det(A_B)} [A_B^{\text{adj}}] b$$

where A_B^{adj} is

[H]

adjugate matrix -

entries are subdeterminants.

- in particular A_B^{adj} integral!
- b, A_B^{adj} integral, $|\det A_B| = 1$

$\Rightarrow X_B$ integral

\Rightarrow every vertex x is integral. \square

Example bipartite matching.

Polytope $P \subseteq \mathbb{R}^{n \times n}$ of "fractional matchings" we used for min-weight-perfect-matching:

Recall:

Let (U, V) be bipartition.

$$\underline{P} := \left\{ X \in \mathbb{R}^{n \times n} \mid \sum_j X_{ij} = 1 \quad \forall i \in U \right.$$

$$\left. \sum_i X_{ij} = 1 \quad \forall j \in V \right.$$

$$x_{ij} \geq 0 \quad \forall i \in U, j \in V. \right\}$$

$$:= \left\{ x : Ax = b, x \geq 0 \right\}.$$

Theorem: The matrix A

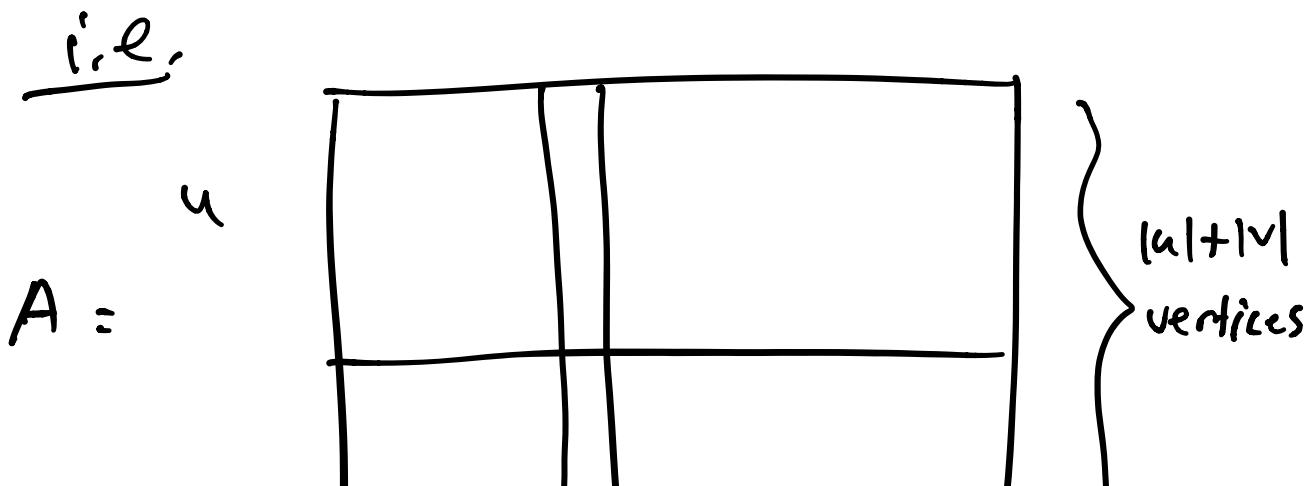
is I II. submodular.

is totally on
(what we
proved in lecture 2 !
).

Cor : $\text{M.W.P.M.} = \min \{ C^T X : X \in P \}$.

Proof : What's A look like?

A^T is incidence matrix of
complete bipartite graph.





④ To show A is TU, consider
square submatrix M & look at
cases:

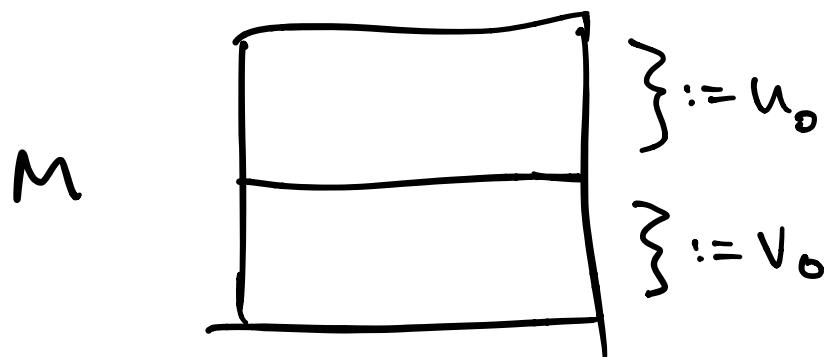
1) if M has 0 row/col,



2) if M has row/col
w/ only one 1,



3) M has ≥ 2 nonzero entries per row & col.



$$\mathbb{I}_{U_0} = \begin{bmatrix} & \\ & \end{bmatrix}_{v_0}^{u_0} \quad \mathbb{I}_{V_0} = \begin{bmatrix} & \\ & \end{bmatrix}_{v_0}^{u_0}$$

(add up rows of M in U_0 ,
get \mathbb{I}^T).

Similarly

\Rightarrow two distinct solns. to $\boxed{\quad}_j$
rows not lin indep.



P is integral if all
its vertices are integral.



$$P = \text{conv}(\text{integer points in } P)$$

L.P $\max c^T x$
 $x \in P$

= I.P $\max c^T x$
 $x \in D$

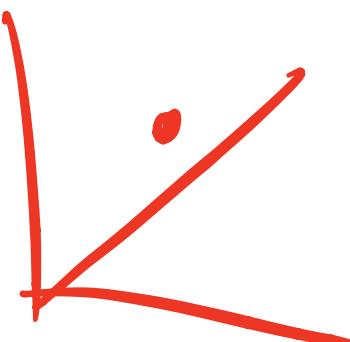
$$x \in \mathbb{R}$$
$$x \in \mathbb{R}^n$$

for all $c \in \mathbb{R}^n$
(maximizers same.)

$$x + y = 1$$

$$x + z = 1$$

$$y + z = 1$$



$$\left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right)$$

IP infeasible

P not integral, IP infeasible,
LP $c^T \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2} \right)$.